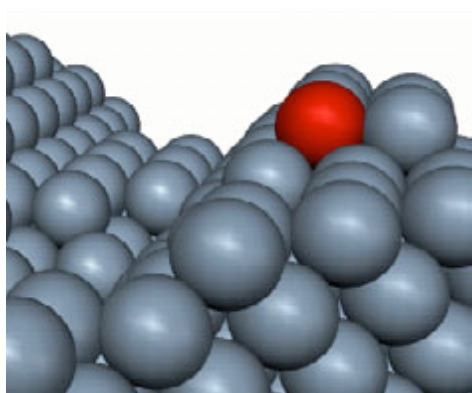


Simulations of kinetic events at the atomic scale

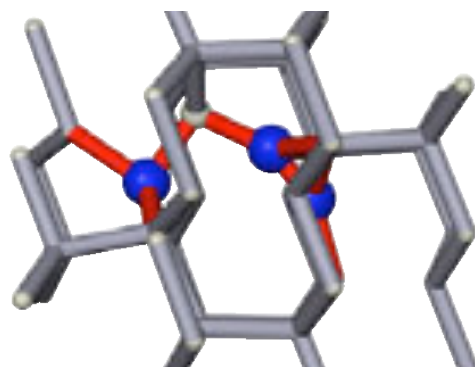
Graeme Henkelman

UT Austin

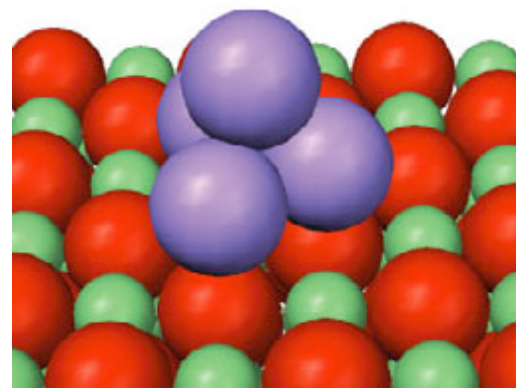
Al / Al(100)



B_3I / Si



Pd / MgO

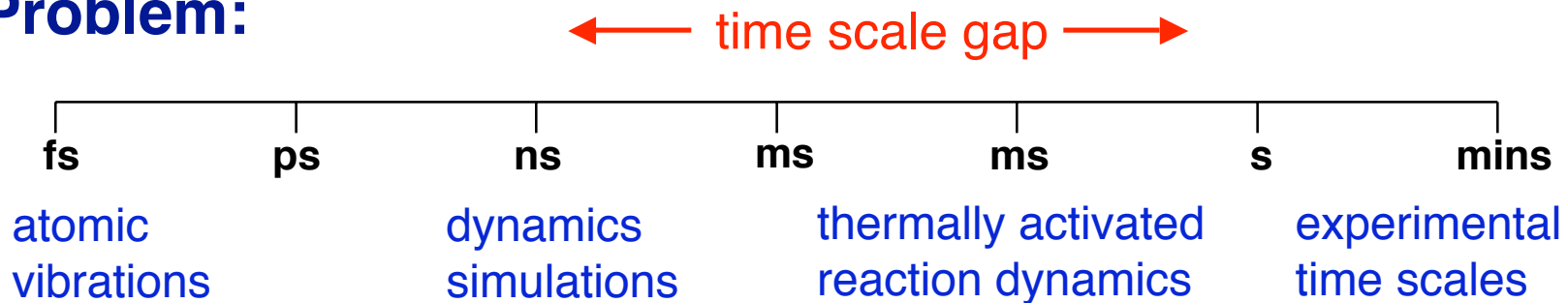


How can we simulate the dynamics of molecular systems over experimental time scales?

Objective:

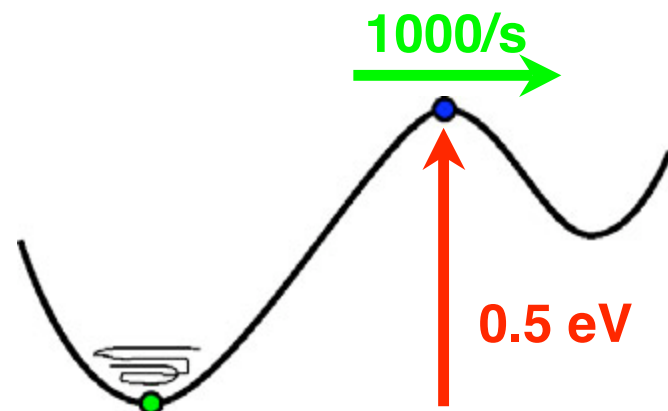
To calculate dynamics of a surface over time scales which are much longer than can be calculated with direct classical dynamics.

Problem:



Most interesting transitions are rare events (much slower than vibrations)

Simulating a transition for a typical rare event with classical dynamics can require $\sim 10^{12}$ force evaluations



Transition state theory

A statistical theory for calculating the rate of slow thermal processes

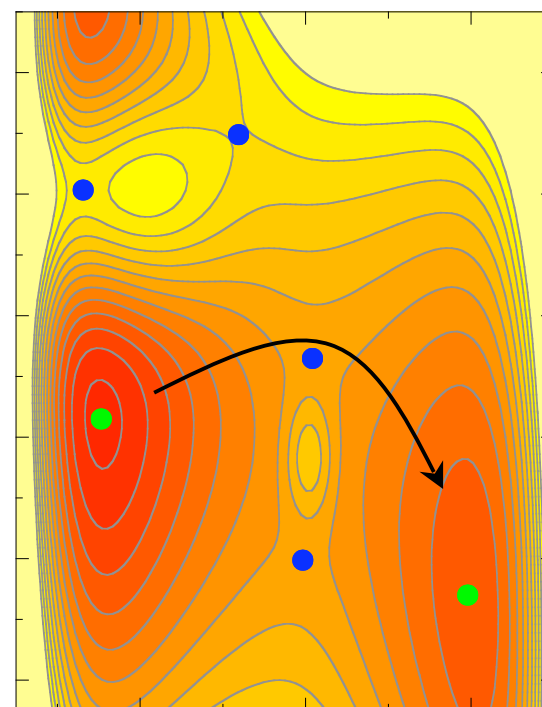
The primary task is to find an N-1 dimensional dividing surface that represents a bottle neck for the transition

Harmonic transition state theory

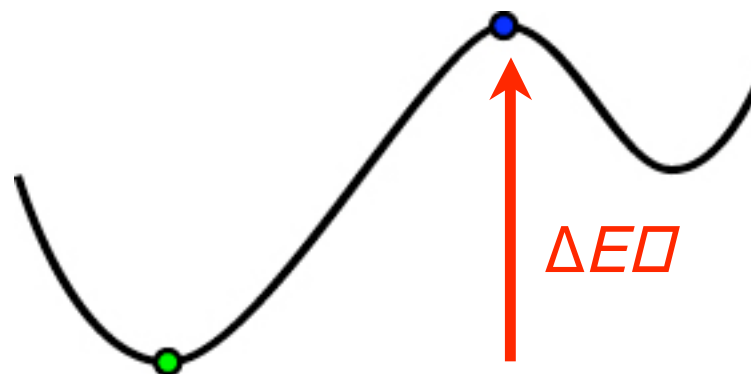
Need to find saddle points on the energy surface

Rate of escape through each saddle point region:

$$Rate = \frac{\prod_{i=1..N} v_i}{\prod_{j=1..N-1} v_j^\ddagger} \cdot e^{-\frac{\Delta E}{k_B T}}$$

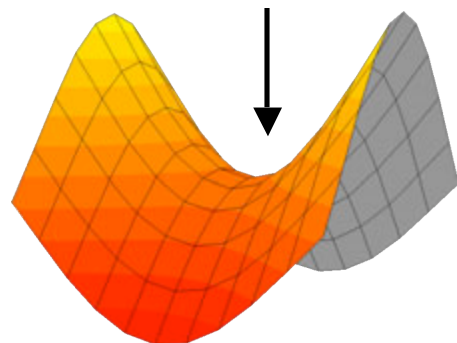


● Minima
● Saddle Point

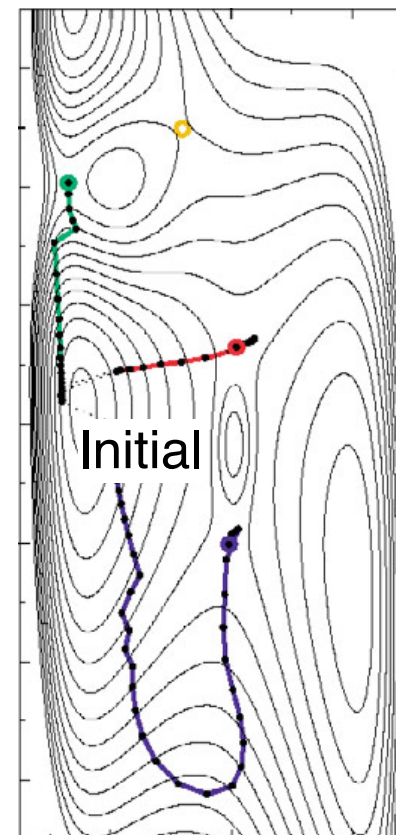
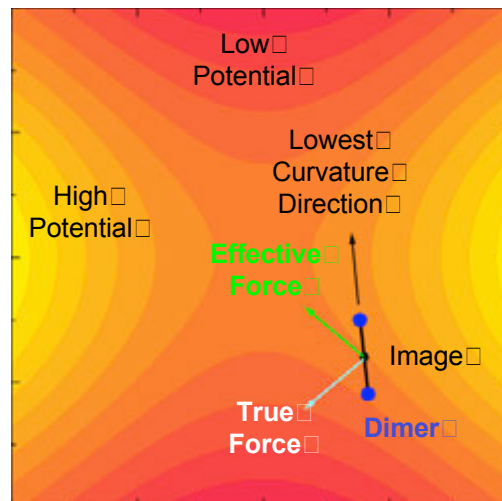


Methods for determining reaction rates

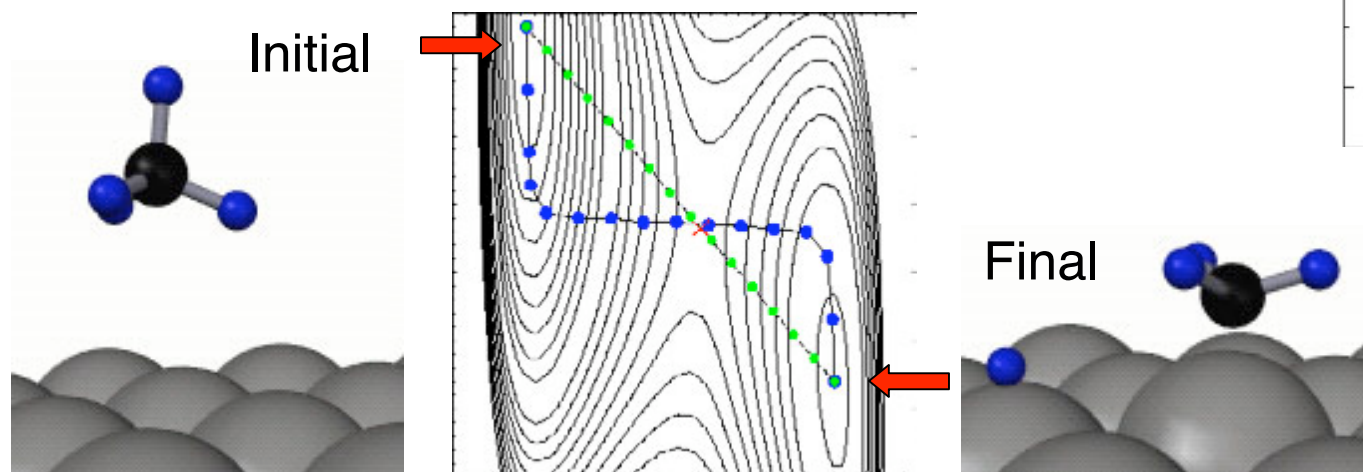
Finding Saddle points



Single Ended:



Double Ended:



The Nudged Elastic Band Method

Recent developments:

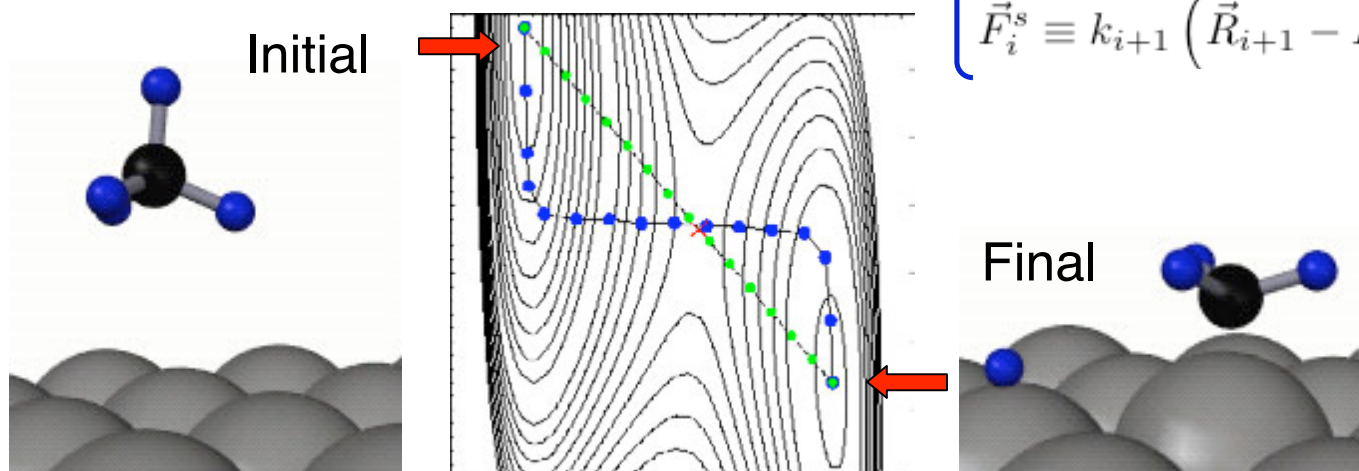
- Improved tangent
- Climbing image
- Double nudging (Wales)
- ☐ Second order optimizers
 - conjugate gradients
 - quasi-newton (bfgs)
- ☐ Internal coordinates
- ☐ Rigid constraints (string method)

Force on each image:

$$\vec{F}_i^{nudged} = -\vec{\nabla}V(\vec{R}_i)|_{\perp} + \vec{F}_i^s \cdot \hat{\tau}_{\parallel} \hat{\tau}_{\parallel} \quad \left\{ \begin{array}{l} \vec{\nabla}V(\vec{R}_i)|_{\perp} = \vec{\nabla}V(\vec{R}_i) - \vec{\nabla}V(\vec{R}_i) \cdot \hat{\tau}_{\parallel} \hat{\tau}_{\parallel} \\ \vec{F}_i^s \equiv k_{i+1} (\vec{R}_{i+1} - \vec{R}_i) - k_i (\vec{R}_i - \vec{R}_{i-1}) \end{array} \right.$$

↪ potential

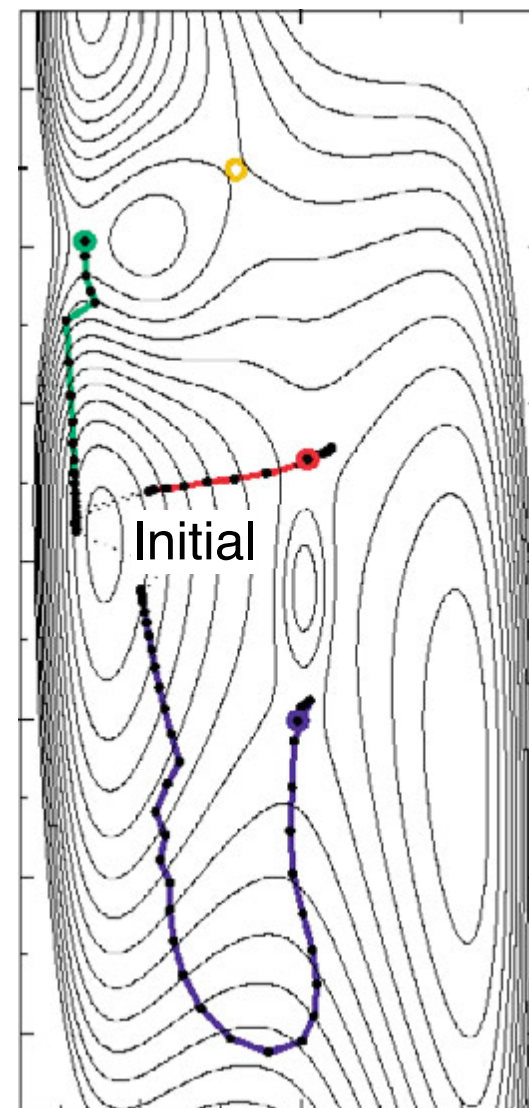
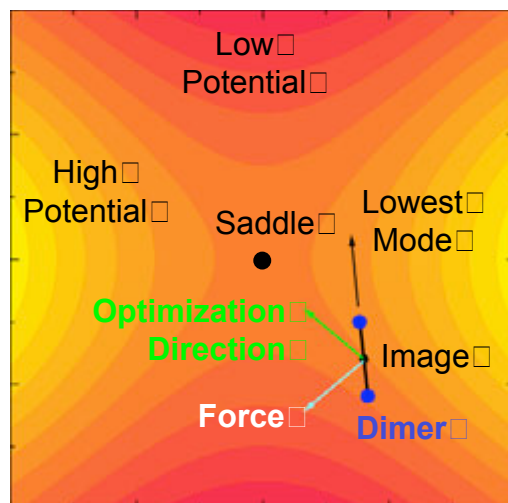
↪ springs



Minimum mode following methods

Dynamics: know the initial state, but don't know the final state!

- Find the lowest curvature mode
 - Dimer method (Voter, Henkelman, Jónsson)
 - Lanczos (Barkema, Mousseau)
 - Langrange multipliers (Wales)
- Follow the minimum mode up the potential, minimize in all other modes
- Many independent searches can be used to find unknown reaction mechanisms

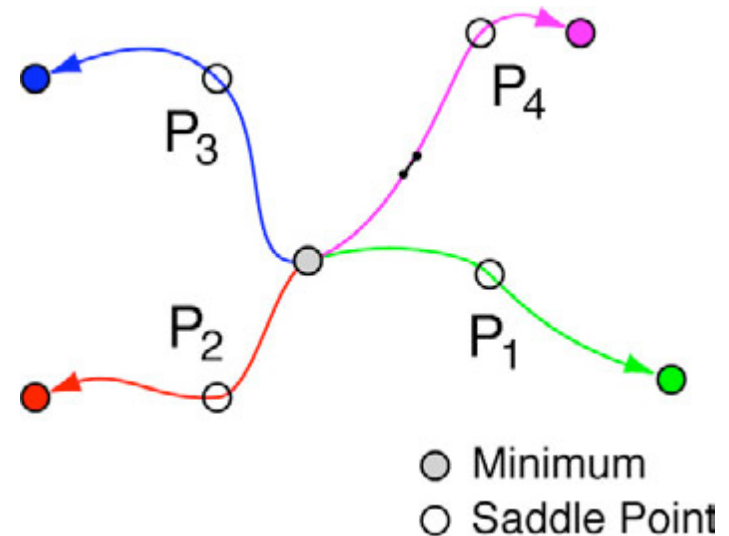


Adaptive kinetic Monte Carlo

Combine saddle point searches with kinetic Monte Carlo (KMC):

Standard KMC

1. Find low energy saddle points using a min-mode method.
2. Choose one processes from a Boltzman distribution.
3. Hop to the final state of the chosen process.
4. Increment time by an average amount Δt .
5. Repeat.



Probability: $p_i \propto r_i$

Rate: $r_i = v e^{-\Delta E_i / k_B T}$

Time: $\overline{\Delta t} = \frac{1}{\sum r_i}$

Al / Al(100) ripening

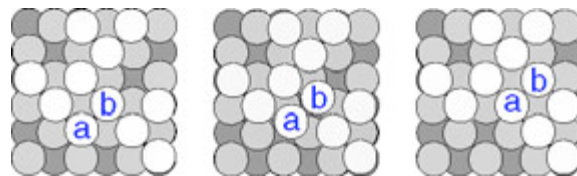
Complex events do happen

- A compact island forms in 1 ms at 300K
- Find many events which are not included in standard KMC event tables

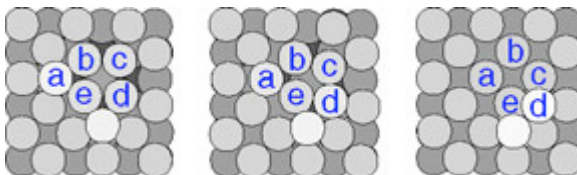
ΔE (eV)

Events

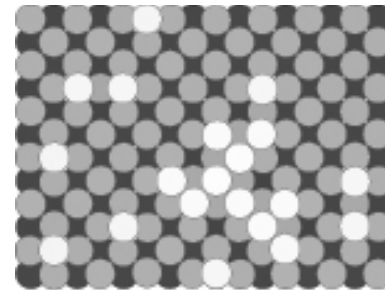
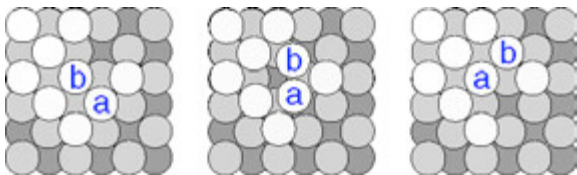
0.16



0.05

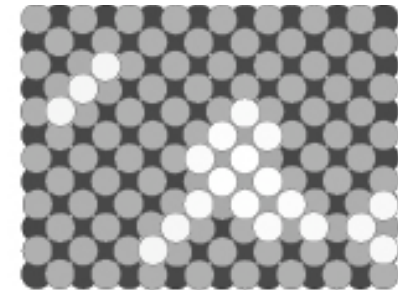


0.24



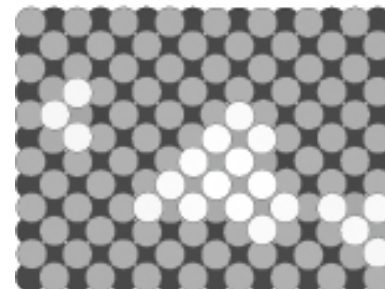
n = 1

t = 0 ns



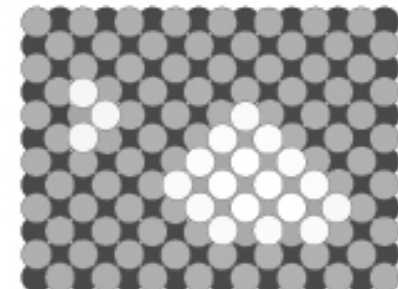
n = 10

t = 6 ns



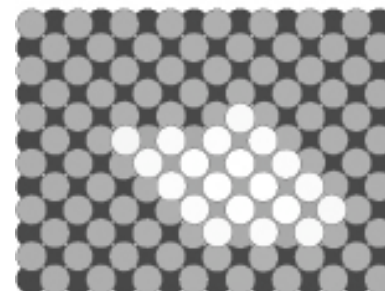
n = 344

t = 70 ns



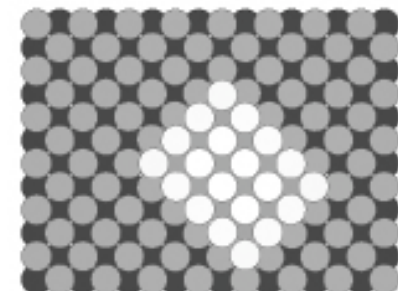
n = 1000

t = 8 μs



n = 7902

t = 10 μs



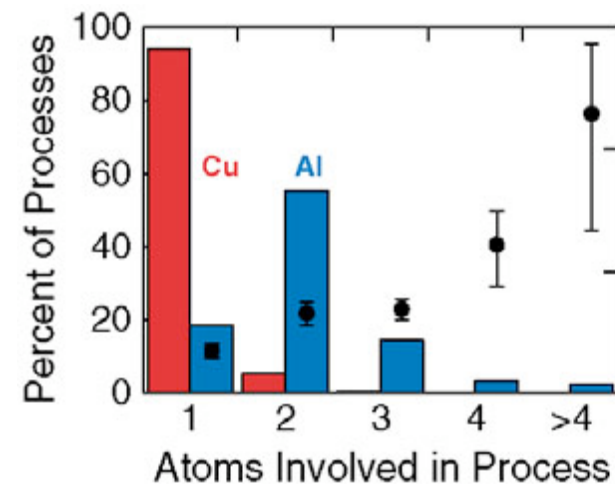
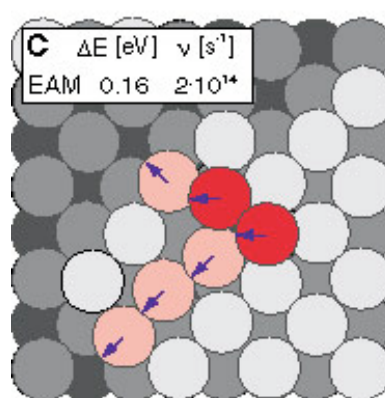
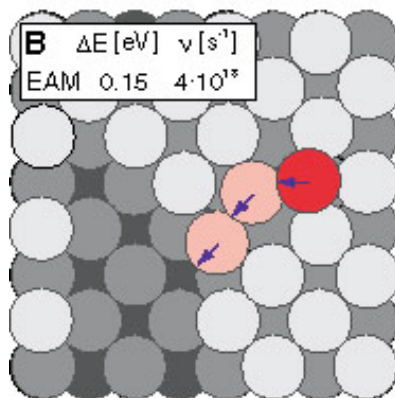
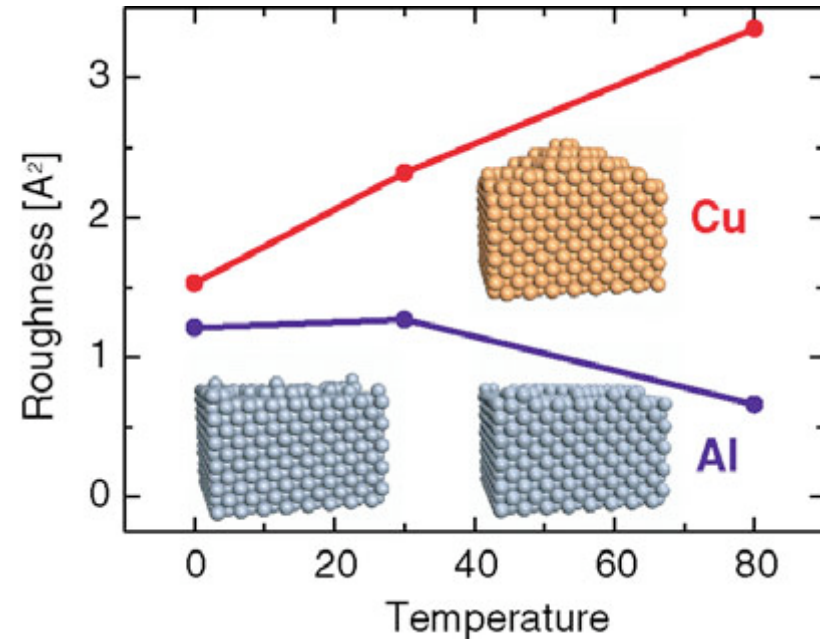
n = 65720

t = 1 ms

Al / Al(100) growth

Multi-atom events can be important for dynamics

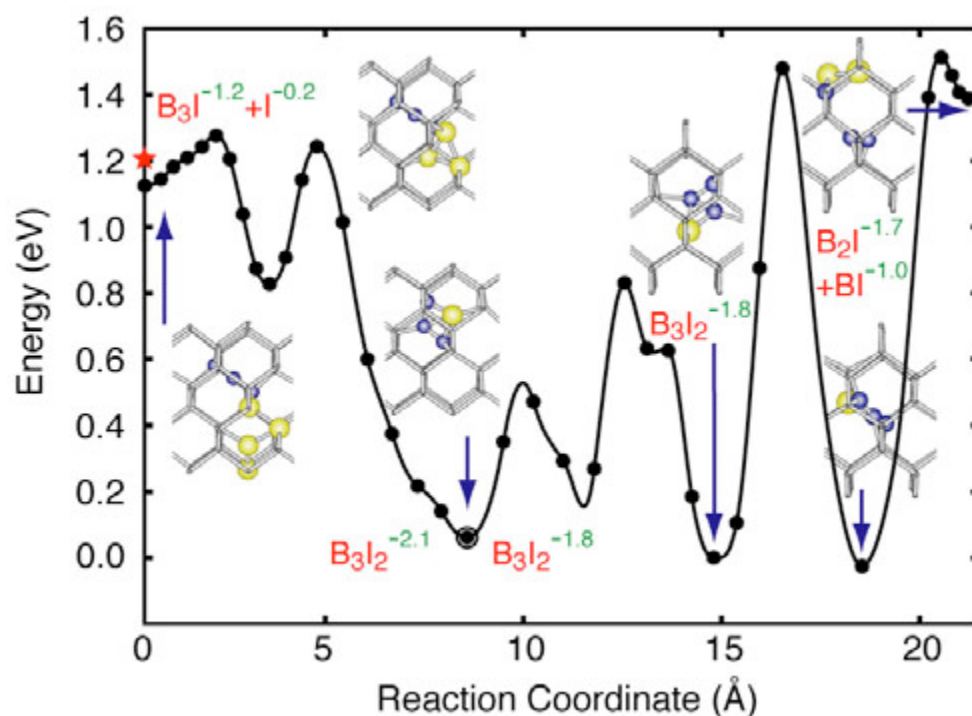
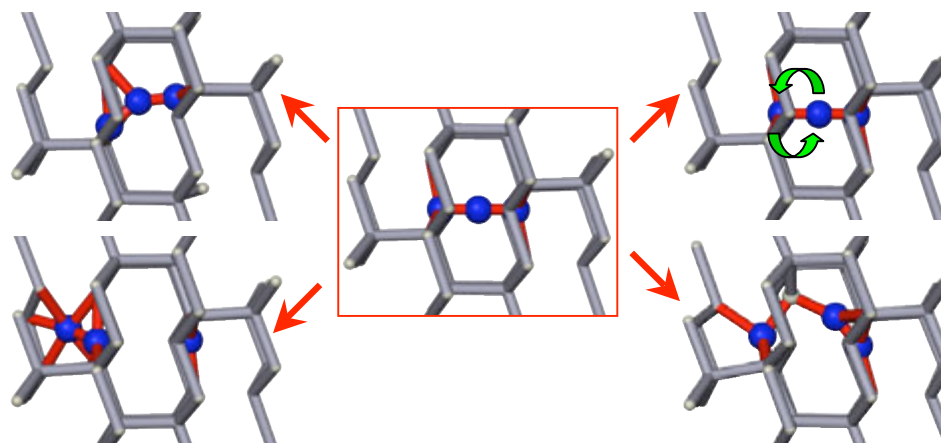
- Exchange events involving more than one atom lead to smooth growth on Al at 77K
- Single atom ripening events contribute to rough growth on Cu



Dynamics from Density Functional Theory

When there is no accurate empirical potential ...

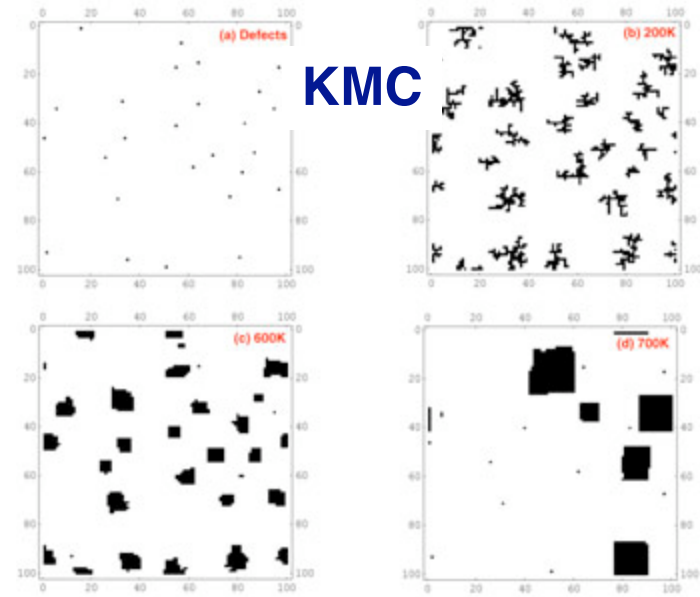
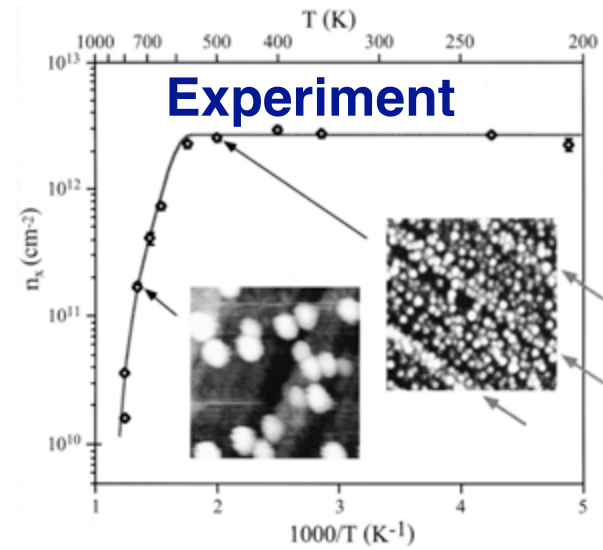
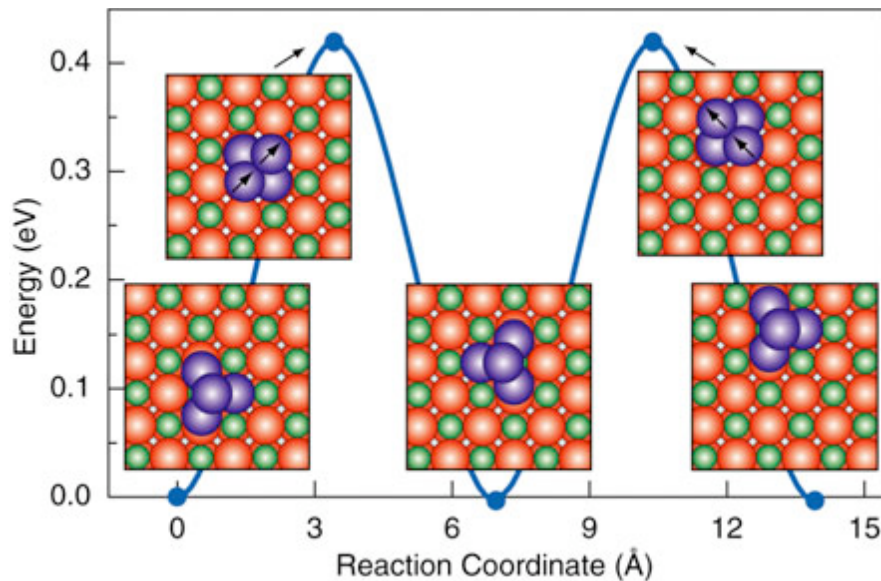
- Saddle points can be found directly with DFT.
- Min-mode following methods are in VASP, SIESTA, SOCORRO, and CASTEP
- Only a few tens of saddle point searches are possible in each new state
- Can be used to find unexpected reaction pathways and dynamics



From DFT to kinetic Monte Carlo: Pd / MgO

If all important mechanisms are found,
KMC can reach longer time scales

- DFT calculations of Pd diffusion on MgO show that small clusters are more mobile than the Pd monomer
- DFT diffusion rates are used in a KMC simulation to compare with experiment

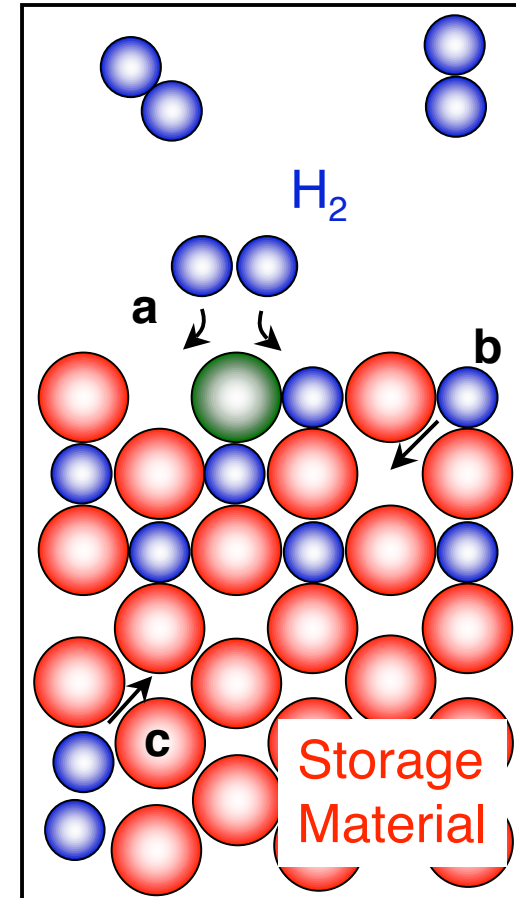


Application to Hydrogen Storage Materials

Challenges for simulating the kinetics of hydrogen storage materials

- a. Need to develop potentials for modeling H_2 dissociative adsorption at catalysts, or use DFT
- b. Since H diffusion barriers will change with environment, it is unlikely that any few reaction mechanisms can be used with standard KMC to model the storage kinetics of a material
- c. A phase change in a material will limit the use of traditional kinetic Monte Carlo

Provides an exciting opportunity for new methods!



Quantum effects for hydrogen kinetics

Zero point and tunneling corrections

- Estimate using classical normal modes **above** T_c

$$T_c = \frac{\hbar |\nu^*|}{k_B}$$

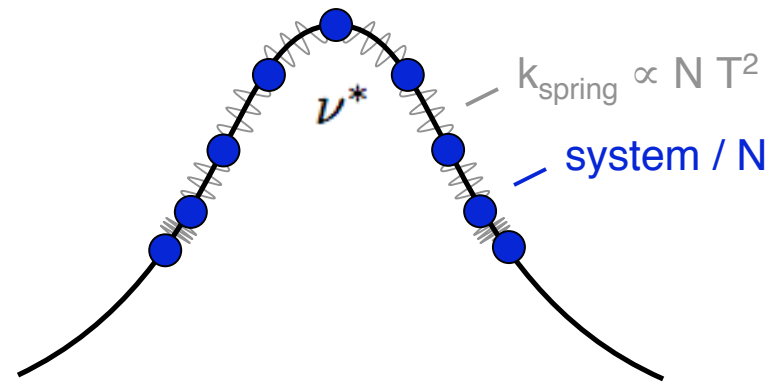
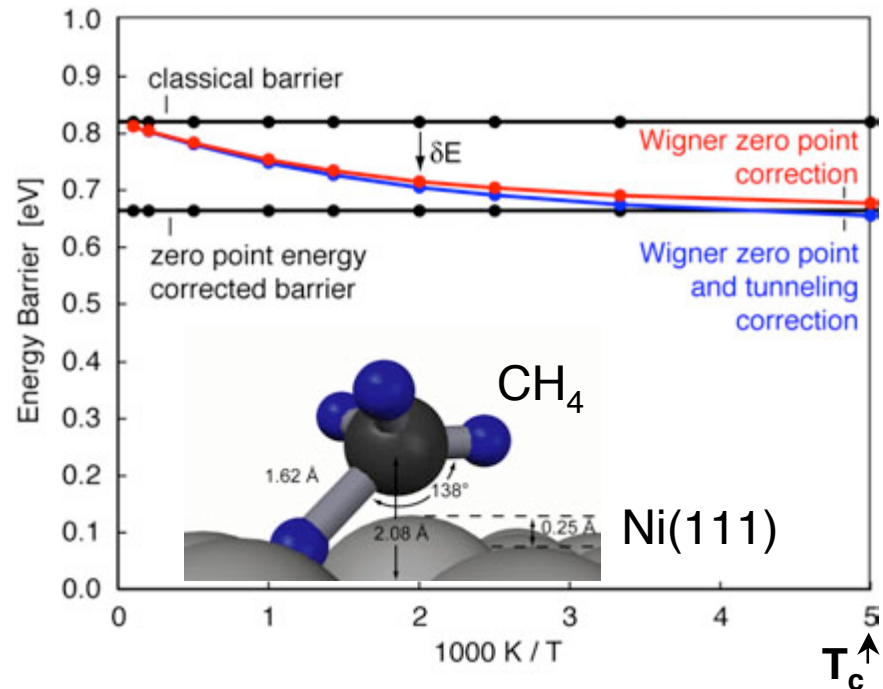
Wigner correction:

$$\delta E_{\text{wig}} = -k_B T \ln \left[\frac{\prod_i \sinh(x_i^{\text{init}})/x_i^{\text{init}}}{\prod_i \sinh(x_i^{\ddagger})/x_i^{\ddagger}} \right]$$

where $x_i = \hbar \nu_i / 2k_B T$

- Below** T_c , find instanton using Min-mode following methods (Jónsson)

Instanton: Saddle point for Feynman chain to cross the barrier



Adaptive kinetic Monte Carlo

Strengths and weaknesses

- Saddle point searches are independent, and can be computed in a parallel (distributed) environment
- Need to find all important (low energy) reaction mechanisms
 - contrast with accelerated dynamics methods (Voter)
- Accuracy determined by sampling:
 - Fewer searches for expensive (DFT) calculations
 - Extensive sampling when using empirical potentials
- No simple relation between sampling and accuracy
- If the important processes are known, they can be used in a KMC simulation to reach longer time and length scales
- Can include quantum effects as necessary

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UT - Austin

Robert A. Welch Foundation

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(State of Texas)

-

Collaborators

UT: Lijun Xu Nathan Froemming
 Wenjie Tang Dan Sheppard

Other

groups: Hannes Jónsson (Iceland)
 Arthur Voter (LANL)
 Blas Uberuaga (LANL)
 Charles Campbell (UW)

Freely available software tools

- <http://eon.cm.utexas.edu/>
- <http://theory.cm.utexas.edu/fida/>
- <http://theory.cm.utexas.edu/vtsttools/>
- <http://theory.cm.utexas.edu/bader/>

The EON distributed computing project

FIDA Distributed computing framework
upon which eon is built

Dimer, NEB, and dynamical matrix
methods implemented in the VASP code

Bader charge density analysis